# Molecular Communication with a Reversible Adsorption Receiver

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Abstract—In this paper, we present an analytical model for a diffusive molecular communication (MC) system with a reversible adsorption receiver in a fluid environment. The time-varying spatial distribution of the information molecules under the reversible adsorption and desorption reaction at the surface of a bio-receiver is analytically characterized. Based on the spatial distribution, we derive the number of newly-adsorbed information molecules expected in any time duration. Importantly, we present a simulation framework for the proposed model that accounts for the diffusion and reversible reaction. Simulation results show the accuracy of our derived expressions, and demonstrate the positive effect of the adsorption rate and the negative effect of the desorption rate on the net number of newly-adsorbed information molecules expected. Moreover, our analytical results simplify to the special case of an absorbing receiver.

## I. INTRODUCTION

Conveying information over a distance has been a problem over decades, and is urgently demanded for different dimensions and various environments. The conventional solution is to utilize electrical- or electromagnetic-enabled communication, which is unfortunately inapplicable or inappropriate in very small dimensions or in specific environments, such as in salt water, tunnels, or human bodies. Recent breakthroughs in bio-nano technology have motivated molecular communication [1] to be a biologically-inspired technique for nanonetworks, where devices with functional components on the scale of 1 to 100 nanometers, namely nanomachines, share information over distance via chemical signals in nanometer to micrometer scale environments.

Diffusion-based MC is the most simple, general and energy efficient transportation paradigm without the need for external energy or infrastructure, where molecules propagate via the random motion, namely Brownian motion, caused by collisions with the fluid's molecules. Examples include deoxyribonucleic acid (DNA) signaling among DNA segments [2] and calcium signaling among cells [3].

In a practical bio-inspired system, the surface of a receiver is covered with selective receptors, which are sensitive to a specific type of information molecule (e.g., specific peptides or calcium ions). The surface of the receiver may adsorb or bind with this specific information molecule [4]. One example is that the influx of calcium towards the center of a receiver (e.g., cell) is induced by the reception of a calcium signal [5]. Despite growing research efforts, the chemical reaction receiver is rarely accurately modeled and characterized in most of the literature except the works from Yilmaz [6–8] and Chou [9], since the local reactions complicate the solution of the reaction-diffusion equations.

Unlike existing works on MC, we consider the reversible adsorption and desorption (A&D) receiver, which is capable of adsorbing a certain type of information molecule near its surface, and desorbing the information molecules previously adsorbed at its surface. A&D is a widely-observed process for colloids [10], proteins [11], and polymers [12]. Also, the A&D process simplifies to the special case of an infinitely absorbing receiver. However, its modeling, analysis, and simulation in the MC domain have never been investigated since the dynamic concentration change near the surface is more challenging than existing works with a passive receiver or an absorbing receiver.

From a theoretical perspective, researchers have derived the equilibrium concentration of A&D [13], which is insufficient to model the time-varying channel impulse response (and ultimately the communications performance) of an A&D receiver. From a simulation perspective, the simulation design for the A&D process of molecules at the surface of a *planar* receiver was proposed in [13]. However, the simulation procedure for the MC communication system with a spherical A&D receiver, where the information molecules, triggered by the transmission of multiple pulses, propagate via free-diffusion through the channel, and contribute to the received signal through A&D at the surface of the receiver, has never been solved and reported. This is due to the complexity in modeling the coupling effect of adsorption and desorption under diffusion, as well as accurately and dynamically tracking the location and the number of diffused molecules, adsorbed molecules and desorbed molecules.

Despite the aforementioned challenges, in this paper we consider the diffusion-based MC system with a point transmitter and an A&D receiver. The goal of this paper is to characterize the impact of the A&D receiver on the net number of newly-adsorbed molecules expected. Our major contributions are summarized as follows.

 We present an analytical model for the diffusion-based MC system with an A&D receiver. We derive the exact expression for the channel impulse response at a spherical A&D receiver in a three dimensional (3D) fluid environment due to a single release of multiple molecules (single transmission). We then derive the *net* number of *newly*-adsorbed molecules expected at the surface of the A&D receiver in any time duration.

2) We propose a simulation algorithm to simulate the diffusion, adsorption and desorption behavior of information molecules based on a particle-based simulation framework. Unlike existing simulation platforms (e.g., Smoldyn [14], NanoNS [15]), our simulation algorithm captures the dynamic process of a MC system, which are the molecule emission, free diffusion, and A&D at the surface of the receiver. Our simulation results are in close agreement with the derived number of adsorbed molecules expected.

The rest of this paper is organized as follows. In Section II, we introduce the system model . In Section III, we present the channel impulse response of information molecules. In Section IV, we present the simulation framework. In Section V, we discuss the numerical and simulation results. In Section VI, we conclude our contributions.

### II. SYSTEM MODEL

We consider a 3-dimensional (3D) diffusion-based MC system in a fluid environment with a point transmitter and a spherical A&D receiver. We assume spherical symmetry where the transmitter is *effectively* a spherical shell and the molecules are released from random points over the shell; the actual angle to the transmitter when a molecule hits the receiver is ignored, so this assumption cannot accommodate a flowing environment. The point transmitter is located at a distance  $r_0$ from the center of the receiver and is at a distance  $d = r_0 - r_r$ from the nearest point on the surface of the receiver with radius  $r_r$ . The extension to an asymmetric spherical model that accounts for the actual angle to the transmitter when a molecule hits the receiver complicates the derivation of the channel impulse response, and may be solved following [16].

We assume all receptors are equivalent and can accommodate at most one adsorbed molecule. The ability of a molecule to adsorb at a given site is independent of the occupation of neighboring receptors. The spherical receiver is assumed to have no physical limitation on the number of molecules adsorbed to the receiver surface (i.e., we ignore saturation). This is an appropriate assumption for a sufficiently low number of adsorbed molecules, or for a sufficiently high concentration of receptors. We also assume perfect synchronization between the transmitter and the receiver as in most literature [6–8]. We consider three processes: emission, propagation, and reception, which are detailed in the following.

## A. Emission

The point transmitter releases one type of information molecule (e.g., hormones, pheromones, or deoxyribonucleic acid (DNA)) to the receiver for information transmission. The transmitter emits  $N_{tx}$  information molecules at t = 0, where we define the initial condition as [17, 3.61]

$$C(r, t \to 0 | r_0) = \frac{1}{4\pi r_0^2} \delta(r - r_0), \qquad (1)$$

where  $C(r, t \to 0 | r_0)$  is the molecule distribution function at time  $t \to 0$  and distance r with initial distance  $r_0$ .

We also define the first boundary condition as

$$\lim_{r \to \infty} C\left(r, t | r_0\right) = 0, \tag{2}$$

such that a molecule that diffuses extremely far away from the receiver is effectively removed from the fluid environment.

#### B. Diffusion

Once the information molecules are emitted, they diffuse by randomly colliding with other molecules in the environment. This random motion is called Brownian motion [2]. The concentration of information molecules is assumed to be sufficiently low that the collisions between those information molecules are ignored [2], such that each information molecule diffuses independently with constant diffusion coefficient D. The propagation model in a 3D environment is described by Fick's second law [2, 7]:

$$\frac{\partial \left( r \cdot C\left( r, t | r_0 \right) \right)}{\partial t} = D \frac{\partial^2 \left( r \cdot C\left( r, t | r_0 \right) \right)}{\partial r^2}, \qquad (3)$$

where the diffusion coefficient is found experimentally [18].

## C. Reception

We consider the reversible A&D receiver, which is capable of counting the net number of newly-adsorbed molecules at the surface of the receiver. Any molecule that hits the receiver surface is either adsorbed to the receiver surface or reflected back into the fluid environment, based on the adsorption rate  $k_1$  (length×time<sup>-1</sup>). The adsorbed molecules either desorb or remain stationary at the surface of receiver, based on the desorption rate  $k_{-1}$  (time<sup>-1</sup>).

At t = 0, there are no information molecules at the receiver surface, so the second initial condition is

$$C(r_r, 0|r_0) = 0$$
, and  $C_a(0|r_0) = 0$ , (4)

where  $C_a(t|r_0)$  is the average concentration of molecules that are adsorbed to the receiver surface at time t.

For the solid-fluid interface located at  $r_r$ , the second boundary condition of the information molecules is [13]

$$D\frac{\partial \left(C\left(r, t | r_{0}\right)\right)}{\partial r}\Big|_{r=r_{r}^{+}} = k_{1}C\left(r_{r}, t | r_{0}\right) - k_{-1}C_{a}\left(t | r_{0}\right),$$
(5)

where  $k_1$  and  $k_{-1}$  are non-zero finite constants. Here, the adsorption rate  $k_1$  is approximately limited to the thermal velocity of potential adsorbents (e.g.,  $k_1 < 7 \times 10^6 \ \mu m/s$  for a 50 kDa protein at 37 °C) [13]; the desorption rate  $k_{-1}$  is typically from  $10^{-4} \ s^{-1}$  and  $10^4 \ s^{-1}$  [19].

The surface concentration  $C_a(t|r_0)$  changes over time as follows:

$$\frac{\partial C_a\left(t \mid r_0\right)}{\partial t} = \left. D \frac{\partial \left(C\left(r, t \mid r_0\right)\right)}{\partial r} \right|_{r=r_r^+},\tag{6}$$

which shows that the change in the adsorbed concentration over time is equal to the flux of diffusion molecules towards the surface. Combining (5) and (6), we write

$$\frac{\partial C_a(t|r_0)}{\partial t} = k_1 C(r_r, t|r_0) - k_{-1} C_a(t|r_0), \quad (7)$$

which is known as the Robin or radiation boundary condition, and shows that the equivalent adsorption rate is proportional to the molecule concentration at the surface.

## **III. RECEIVER OBSERVATIONS**

In this section, we first derive the spherically-symmetric spatial distribution  $C(r, t | r_0)$ , which is the probability of finding a molecule at distance r and time t. We then derive the flux at the surface of the A&D receiver, from which we derive the exact number of adsorbed molecules expected at the surface of the receiver. In the following theorem, we solve the time-varying spatial distribution of information molecules at the surface of the receiver.

**Theorem 1.** The expected time-varying spatial distribution of an information molecule released into a 3D fluid environment with a reversible adsorbing receiver is given by

$$C(r, t|r_0) = \frac{1}{4\pi r_0 r \sqrt{4\pi Dt}} \exp\left\{-\frac{(r-r_0)^2}{4Dt}\right\} + \frac{1}{4\pi r_0 r \sqrt{4\pi Dt}} \exp\left\{-\frac{(r+r_0-2r_r)^2}{4Dt}\right\} - \frac{1}{2\pi r} \int_0^\infty \left(e^{-jwt} \varphi_Z^*(w) + e^{jwt} \varphi_Z(w)\right) dw,$$
(8)

where

$$\varphi_Z(w) = Z(jw) = \frac{2\left(\frac{1}{r_r} + \frac{k_1jw}{D(jw+k_{-1})}\right)}{\left(\frac{1}{r_r} + \frac{k_1jw}{D(jw+k_{-1})} + \sqrt{\frac{jw}{D}}\right)} \times \frac{1}{4\pi r_0\sqrt{4Djw}} \exp\left\{-\left(r + r_0 - 2r_r\right)\sqrt{\frac{jw}{D}}\right\}.$$
(9)

and  $\varphi_Z^*(w)$  is the complex conjugate of  $\varphi_Z(w)$ .

Proof: See Appendix A.

We observe that (8) reduces to the *absorbing* receiver [17, Eq. (3.99)] when there is no desorption (i.e.,  $k_{-1} = 0$ ).

To characterize the number of information molecules adsorbed at the surface of the receiver using  $C(r,t|r_0)$ , we define the rate of the coupled reaction (i.e., adsorption and desorption) at the surface of the reversible adsorbing receiver as [17, Eq. (3.106)]

$$K(t|r_0) = 4\pi r_r^2 D \left. \frac{\partial C(r,t|r_0)}{\partial r} \right|_{r=r_r}.$$
 (10)

**Corollary 1.** The rate of the coupling reaction at the surface of a reversible adsorbing receiver is given by

$$K(t|r_0) = 2r_r D \int_0^\infty e^{-jwt} \left[ \sqrt{\frac{jw}{D}} \varphi_Z(w) \right]^* dw + 2r_r D \int_0^\infty e^{jwt} \left[ \sqrt{\frac{jw}{D}} \varphi_Z(w) \right] dw, \quad (11)$$

where  $\varphi_Z(w)$  is as given in (9).

*Proof:* By substituting (8) into (10), we derive the coupling reaction rate at the surface of an A&D receiver as (11).

From **Corollary 1**, we can derive the net change in the number of adsorbed molecules expected for any time interval in the following theorem.

**Theorem 2.** The net change in the number of adsorbed molecules expected at the surface of the receiver during the interval  $[T, T+T_s]$  is derived as

$$\mathbb{E}\left[N_{A\&D}\left(\Omega_{r_r}, T, T+T_s \middle| r_0\right)\right] = 2r_r N_{tx} D$$

$$\times \left[\int_0^\infty \frac{e^{-jwT} - e^{-jw(T+T_s)}}{jw} \left[\sqrt{\frac{jw}{D}}\varphi_Z\left(w\right)\right]^* dw$$

$$+ \int_0^\infty \frac{e^{jw(T+T_s)} - e^{jwT}}{jw} \left[\sqrt{\frac{jw}{D}}\varphi_Z\left(w\right)\right] dw \right], \quad (12)$$

where  $\varphi_Z(w)$  is given in (9),  $T_s$  is the sampling time, and  $\Omega_{r_r}$  represents the spherical receiver with radius  $r_r$ .

*Proof:* The cumulative fraction of particles that are adsorbed at the surface of the receiver at time T is expressed as

$$R_{A\&D}(\Omega_{r_r}, T | r_0) = \int_0^T K(t | r_0) dt.$$
 (13)

Based on (13), the net change of adsorbed molecules expected at the surface of the receiver during the interval  $[T, T+T_s]$  is defined as

$$\mathbb{E} \left[ N_{A\&D} \left( \Omega_{r_r}, T, T + T_s | r_0 \right) \right] = N_{tx} R_{A\&D} \left( \Omega_{r_r}, T + T_s | r_0 \right) - N_{tx} R_{A\&D} \left( \Omega_{r_r}, T | r_0 \right).$$
(14)

Substituting (13) into (14), we derive the expected net change of adsorbed molecules during any observation interval as (12).

#### **IV. SIMULATION FRAMEWORK**

This section describes the stochastic simulation framework of the point-to-point MC system with the A&D receiver described by (5). To accurately capture the locations of individual information molecules, we adopt a particle-based simulation framework with a spatial resolution on the scale of several nanometers [13]. Algorithm 1 Simulation of a MC System with an A&D Receiver Require:  $N_{\text{tx}}$ ,  $r_0$ ,  $r_r$ ,  $\Omega_{r_r}$ , D,  $\Delta t$ ,  $T_s$ ,  $T_b$ ,  $N_{\text{th}}$ 1: procedure INITIALIZATION Determine Simulation End Time 2: Add  $N_{tx}$  emitted molecules 3: 4: For all Simulation Time Step do For all free molecules in environment do 5: 6: Propagate free molecules following  $\mathcal{N}(0, 2D\Delta t)$ Evaluate distance  $d_m$  of molecule to receiver 7: 8: if  $d_m < r_r$  then Update state & location of collided molecule 9: 10: Update # of collided molecules  $N_C$ For all  $N_C$  collided molecules do 11: 12: if Adsorption Occurs then Update # of *newly*-adsorbed molecules  $N_A$ 13: Calculate adsorbed molecule location 14:  $(x_m^A, y_m^A, z_m^A)$ 15: else 16: Reflect the molecule off receiver surface to 17:  $(x_m^{Bo}, y_m^{Bo}, z_m^{Bo})$ 18: For all previously-adsorbed molecules do 19: if Desorption Occurs then 20: Update state & location of desorbed molecule 21: Update # of *newly*-desorbed molecules  $N_D$ 22. 23: Displace *newly*-desorbed molecule to  $(x_m^D, y_m^D, z_m^D)$ 24: Calculate net number of newly-adsorbed molecules, 25: which is  $N_A - N_D$ 26:

# A. Algorithm

We present the algorithm for simulating the MC system with an A&D receiver in Algorithm 1. In the following subsections, we describe the details of Algorithm 1.

#### B. Emission and Diffusion

At time t = 0,  $N_{tx}$  molecules are emitted from the point transmitter at a distance  $r_0$  from the center of the receiver. The time is divided into small simulation intervals of size  $\Delta t$ , and each time instant is represented by  $t_m = m\Delta t$ , where m is the current simulation index. The displacement  $\Delta S$  of a molecule in a 3D fluid environment in one simulation step  $\Delta t$  is modeled as

$$\Delta S = \{ \mathcal{N}(0, 2D\Delta t), \ \mathcal{N}(0, 2D\Delta t), \ \mathcal{N}(0, 2D\Delta t) \}, \ (15)$$

where  $\mathcal{N}(0, 2D\Delta t)$  is the normal distribution. In each simulation step, the number of molecules and their locations are stored.

## C. Adsorption or Reflection

According to the second boundary condition in (6), molecules that collide with the receiver surface are either adsorbed or reflected back. The  $N_C$  collided molecules are identified by calculating the distance between each molecule

and the center of the receiver. Among the collided molecules, the probability of a molecule being adsorbed to the receiver surface, i.e., the adsorption probability, is a function of the diffusion coefficient, which is given as [20, Eq. (10)]

$$P_A = k_1 \sqrt{\frac{\pi \Delta t}{D}}.$$
 (16)

The probability that a collided molecule bounces off of the receiver is  $1 - P_A$ .

It is known that adsorption may occur during the simulation step  $\Delta t$ , and determining exactly where a molecule adsorbed to the surface of the receiver during  $\Delta t$  is a non-trivial problem. To simplify this, we assume that the adsorbed location of a molecule during  $[t_{m-1}, t_m]$  is equal to the location where the line, formed by this molecule's location at the start of the current simulation step  $(x_{m-1}, y_{m-1}, z_{m-1})$  and this molecule's location at the end of the current simulation step after diffusion  $(x_m, y_m, z_m)$ , intersects the surface of the receiver. Assuming that the location of the center of receiver is  $(x_r, y_r, z_r)$ , then the location of the intersection point between this 3D line segment, and a sphere with center at  $(x_r, y_r, z_r)$ in the *m*th simulation step, can be shown to be

$$x_m^A = x_{m-1} + \frac{x_m - x_{m-1}}{\Delta}g,$$
 (17)

$$y_m^A = y_{m-1} + \frac{y_m - y_{m-1}}{\Delta}g,$$
 (18)

$$z_m^A = z_{m-1} + \frac{z_m - z_{m-1}}{\Delta}g,$$
 (19)

where

$$\Delta = \sqrt{\left(x_m - x_{m-1}\right)^2 + \left(y_m - y_{m-1}\right)^2 + \left(z_m - z_{m-1}\right)^2},$$
(20)

$$y = \frac{-b - \sqrt{b^2 - 4ac}}{2a}.$$
 (21)

In (21), we have

(

$$a = \left(\frac{x_m - x_{m-1}}{\Delta}\right)^2 + \left(\frac{y_m - y_{m-1}}{\Delta}\right)^2 + \left(\frac{z_m - z_{m-1}}{\Delta}\right)^2,$$
  

$$b = 2\frac{(x_m - x_{m-1})(x_{m-1} - x_r)}{\Delta} + 2\frac{(y_m - y_{m-1})(y_{m-1} - y_r)}{\Delta} + 2\frac{(z_m - z_{m-1})(z_{m-1} - z_r)}{\Delta},$$
(22)

$$c = (x_{m-1} - x_r)^2 + (y_{m-1} - y_r)^2 + (z_{m-1} - z_r)^2 - r_r^2,$$
(23)

where  $\Delta$  is given in (20).

Of course, due to symmetry, the location of the adsorption site does not impact the overall accuracy of the simulation.

If a molecule fails to adsorb to the receiver, then in the reflection process we make the approximation that the molecule bounces back to its position at the start of the current simulation step. Thus, the location of the molecule

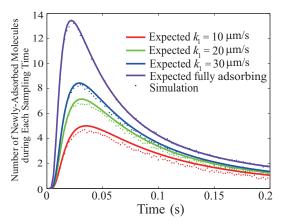


Fig. 1. The net number of newly-adsorbed molecules for various adsorption rates with parameters:  $k_{-1} = 5 \ s^{-1}$ ,  $N_{\rm tx} = 1000$ ,  $r_r = 10 \ \mu{\rm m}$ ,  $d = 1 \ \mu{\rm m}$ ,  $D = 8 \ \mu{\rm m}^2/{\rm s}$ ,  $T_s = 0.002$  s, and the simulation step  $\Delta t = 10^{-5}$  s.

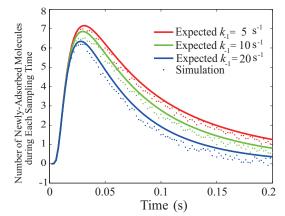


Fig. 2. The net number of newly-adsorbed information molecules for various desorption rates with parameters:  $k_1 = 20 \ \mu m/s$ ,  $N_{\rm tx} = 1000$ ,  $r_r = 10 \ \mu m$ ,  $d = 1 \ \mu m$ ,  $D = 8 \ \mu m^2/s$ ,  $\Delta t = 10^{-4}$  s, and  $T_s = 0.002$  s.

after reflection by the receiver in the mth simulation step is approximated as

$$(x_m^{Bo}, y_m^{Bo}, z_m^{Bo}) = (x_{m-1}, y_{m-1}, z_{m-1}).$$
 (24)

Note that the approximations for molecule locations in the adsorption process and the reflection process can be accurate for sufficiently small simulation steps (e.g.,  $\Delta t < 10^{-7}$  s for the system that we simulate in Section V), but small simulation steps result in poor computational efficiency.

## D. Desorption

In the desorption process, the molecules adsorbed at the receiver boundary either desorb or remain adsorbed. The desorption process can be modeled as a first-order chemical reaction. Thus, the desorption probability of a molecule at the receiver surface during  $\Delta t$  is given by [13, Eq. (22)]

$$P_D = 1 - e^{-k_{-1}\Delta t}.$$
 (25)

The displacement of a molecule after desorption is an important factor for accurate modeling of molecule behaviour. If the simulation step were small, then we might place the desorbed molecule near the receiver surface; otherwise, doing so may result in an artificially higher chance of re-adsorption in the following time step, resulting in an inexact concentration profile. To avoid this, we take into account the diffusion *after* desorption, and place the desorbed molecule away from the surface with displacement  $(\Delta x, \Delta y, \Delta z)$ 

$$(\Delta x, \Delta y, \Delta z) = (f(P_1), f(P_2), f(P_3)),$$
 (26)

where each component was empirically found to be [13, Eq. (27)]

$$f(P) = \sqrt{2D\Delta t} \frac{0.571825P - 0.552246P^2}{1 - 1.53908P + 0.546424P^2}.$$
 (27)

In (26),  $P_1$ ,  $P_2$  and  $P_3$  are uniform random numbers between 0 and 1. Placing the desorbed molecule at a random distance away along the line from the center of the receiver to where the molecule was adsorbed is not sufficiently accurate due to the lack of consideration for the coupling effect of A&D and the diffusion coefficient in (27).

Different from [13], we have the spherical receiver such that the molecule after desorption in our model should be displaced differently. We assume that the location of a molecule after desorption  $(x_m^D, y_m^D, z_m^D)$ , based on its location at the start of the current simulation step and the location of the center of the receiver  $(x_r, y_r, z_r)$ , can be approximated as

$$\begin{aligned} x_m^D &= x_{m-1}^A + \operatorname{sgn} \left( x_{m-1}^A - x_r \right) \Delta x, \\ y_m^D &= y_{m-1}^A + \operatorname{sgn} \left( y_{m-1}^A - y_r \right) \Delta y, \\ z_m^D &= z_{m-1}^A + \operatorname{sgn} \left( z_{m-1}^A - z_r \right) \Delta z. \end{aligned}$$
(28)

In (28),  $\Delta x$ ,  $\Delta y$ , and  $\Delta z$  are given in (26), and sgn ( $\cdot$ ) is the Sign function.

#### E. Reception

The receiver is capable of counting the net change in the number of adsorbed molecules in each simulation step.

#### V. NUMERICAL RESULTS

Fig. 1 and Fig. 2 plot the net change of adsorbed molecules at the surface of the A&D receiver at each sampling time  $T_s$  due to a single bit transmission. The expected analytical curves are plotted using the exact result in (12). The simulation points are plotted by measuring the net change of adsorbed molecules during  $[t, t + T_s]$  using Algorithm 1 described in Section IV, where  $t = nT_s$ , and  $n \in \{1, 2, 3, \ldots\}$ . In both figures, we average the number of newly-adsorbed molecules expected over 1000 independent emissions of  $N_{tx}$  information molecules. We see that the expected number of newlyadsorbed molecules measured using simulation is close to the exact analytical curves. Note that the small gap between the curves results from the local approximations in the adsorption, reflection, and desorption processes in (16)-(19), (24), and (28), which can be reduced by setting smaller simulation step. Fig. 1 examines the impact of the adsorption rate on the net number of newly-adsorbed molecules expected at the surface of the receiver. We fix the desorption rate to be  $k_{-1} = 5 \text{ s}^{-1}$ . The number of newly-adsorbed molecules expected increases with increasing adsorption rate  $k_1$ , as predicted by (5). Compared with the full adsorption receiver (e.i.,  $k_1 = \infty$ ), the A&D receiver has a weaker observed signal. Fig. 2 shows the impact of the desorption rate on the number of newlyadsorbed molecules expected at the surface of the receiver. We set  $k_1 = 20 \ \mu\text{m/s}$ . The number of newly-adsorbed molecules expected decreases with increasing desorption rate  $k_{-1}$ , which is as predicted by (5).

At the receiver side, the number of newly-adsorbed molecules during each symbol interval could be compared with a threshold to demodulate the signal. From the communication point of view, Fig. 1 shows that the higher adsorption rate makes the received signal more distinguishable. In Fig. 1 and Fig. 2, the shorter tail due to the lower adsorption rate and the higher desorption rate corresponds to less intersymbol interference.

## VI. CONCLUSION

In this paper, we modeled the diffusion-based MC system with the A&D receiver. We derived the exact expression for the net number of newly-adsorbed information molecules expected at the surface of the receiver. We also presented a simulation algorithm that captures the behavior of each information molecule with the stochastic reversible reaction at the receiver. We revealed that the number of newly-adsorbed information molecules expected at the surface of the receiver increases with increasing adsorption rate and with decreasing desorption rate. Our ongoing work is comparing our proposed model with existing receiver models and considering the impact on bit error performance. Our analytical model and simulation framework provide a foundation for the accurate design and analysis of a more complex and realistic receiver in molecular communication.

# APPENDIX A Proof of Theorem 1

We first partition the spherically symmetric distribution into two parts using the method applied in [17]

$$r \cdot C(r, t | r_0) = r \cdot g(r, t | r_0) + r \cdot h(r, t | r_0), \qquad (29)$$

where

$$g(r, t \to 0 | r_0) = \frac{1}{4\pi r_0} \delta(r - r_0), \qquad (30)$$

$$h(r, t \to 0 | r_0) = 0.$$
 (31)

Then, by substituting (29) into (3), we have

$$\frac{\partial \left(r \cdot g\left(r, t \mid r_{0}\right)\right)}{\partial t} = D \frac{\partial^{2} \left(r \cdot g\left(r, t \mid r_{0}\right)\right)}{\partial r^{2}}, \qquad (32)$$

and

$$\frac{\partial \left(r \cdot h\left(r, t | r_0\right)\right)}{\partial t} = D \frac{\partial^2 \left(r \cdot h\left(r, t | r_0\right)\right)}{\partial r^2}.$$
 (33)

To derive  $g(r, t | r_0)$ , we perform a Fourier transformation on  $rg(r, t | r_0)$  to yield

$$G(k, t | r_0) = \int_{-\infty}^{\infty} rg(r, t | r_0) e^{-ikr} dr, \qquad (34)$$

and

$$r \cdot g(r, t | r_0) = \frac{1}{2\pi} \int_{-\infty}^{\infty} G(k, t | r_0) e^{ikr} dk.$$
 (35)

We then perform the Fourier transformation on (32) to yield

$$\frac{dG(k,t|r_0)}{dt} = -Dk^2 G(k,t|r_0).$$
 (36)

According to (36) and the uniqueness of the Fourier transform, we derive

$$G(k, t|r_0) = K_g \exp\{-Dk^2 t\},$$
(37)

where  $K_g$  is an undetermined constant.

The Fourier transformation performed on (30) yields

$$G(r, t \to 0 | r_0) = \frac{1}{4\pi r_0} e^{-ikr_0}.$$
 (38)

Combining (37) and (38), we arrive at

$$G(k, t|r_0) = \frac{1}{4\pi r_0} e^{-ikr_0} \exp\left\{-Dk^2t\right\}.$$
 (39)

Substituting (39) into (35), we find that

$$r \cdot g(r, t | r_0) = \frac{1}{8\pi r_0 \sqrt{\pi Dt}} \exp\left\{-\frac{(r - r_0)^2}{4Dt}\right\}.$$
 (40)

By performing the Laplace transform on (40), we write

$$\mathcal{L}\left\{r \cdot g\left(r, t | r_{0}\right)\right\} = \frac{1}{4\pi r_{0}\sqrt{4Ds}} \exp\left\{-\left|r - r_{0}\right| \sqrt{\frac{s}{D}}\right\}.$$
(41)

We then focus on solving the solution  $h(k, t|r_0)$  by first performing the Laplace transform on  $h(k, t|r_0)$  and (33) as

$$H(r, s|r_0) = \mathcal{L} \{ h(r, t|r_0) \} = \int_0^\infty h(r, t|r_0) e^{-s\tau} d\tau,$$
(42)

and

$$srH\left(r,s|r_{0}\right) = D\frac{\partial^{2}\left(rH\left(r,s|r_{0}\right)\right)}{\partial r^{2}},$$
(43)

respectively.

According to (43), the Laplace transform of the solution with respect to the boundary condition in (43) is

$$rH(r, s|r_0) = f(s) \exp\left\{-\sqrt{\frac{s}{D}}r\right\},$$
(44)

where f(s) needs to satisfy the second initial condition in (4), and the second boundary condition in (5) and (6).

Having the Laplace transform of  $\{r \cdot g(r, t | r_0)\}$  and  $h(r, t | r_0)$  in (41) and (44), and performing a Laplace transformation on (29), we derive

$$\hat{C}(r, s | r_0) = G(r, s | r_0) + rH(r, s | r_0) 
= \frac{1}{8\pi r_0 \sqrt{Ds}} \exp\left\{-|r - r_0| \sqrt{\frac{s}{D}}\right\} + f(s) \exp\left\{-\sqrt{\frac{s}{D}}r\right\} 
(45)$$

where  $\tilde{C}\left(r, \left.s\right|r_{0}\right) = \int_{0}^{\infty} C\left(r, \left.t\right|r_{0}\right) e^{-st} dt.$ 

To solve f(s), we perform the Laplace transform on the Robin boundary condition in (7) to yield

$$\tilde{C}_{a}(s|r_{0}) = \frac{k_{1}\tilde{C}(r_{r},s|r_{0})}{s+k_{-1}},$$
(46)

where  $\tilde{C}_{a}(r, s | r_{0}) = \int_{0}^{\infty} C_{a}(r, t | r_{0}) e^{-st} dt$ .

We then perform the Laplace transform on the second initial condition in (4) and the second boundary condition in (5) as

$$D\frac{\partial \left(\tilde{C}\left(r,t|r_{0}\right)\right)}{\partial r}\bigg|_{r=r_{r}} = k_{1}\tilde{C}\left(r_{r},s|r_{0}\right) - k_{-1}\tilde{C}_{a}\left(s|r_{0}\right).$$
(47)

Substituting (38) into (47), we obtain

$$D\frac{\partial\left(\tilde{C}\left(r,t|r_{0}\right)\right)}{\partial r}\bigg|_{r=r_{r}} = \frac{k_{1}s}{s+k_{-1}}\tilde{C}\left(r_{r},s|r_{0}\right).$$
 (48)

To facilitate the analysis, we express the Laplace transform on the second boundary condition as

$$\frac{\partial \left( r \cdot \tilde{C}\left(r, s \mid r_{0}\right) \right)}{\partial r} \bigg|_{r=r_{r}} = \left( 1 + \frac{r_{r}k_{1}s}{D\left(s+k_{-1}\right)} \right) \tilde{C}\left(r, s \mid r_{0}\right).$$
(49)

Substituting (45) into (49), we determine f(s) as

$$f(s) = \frac{\left(\sqrt{\frac{s}{D}} - \frac{1}{r_r} - \frac{k_{1s}}{D(s+k_{-1})}\right)}{\left(\sqrt{\frac{s}{D}} + \frac{1}{r_r} + \frac{k_{1s}}{D(s+k_{-1})}\right)} \frac{\exp\left\{-\left(r_0 - 2r_r\right)\sqrt{\frac{s}{D}}\right\}}{4\pi r_0\sqrt{4Ds}}.$$
(50)

Having (45) and (50), and performing the Laplace transform of the concentration distribution, we derive

$$r\tilde{C}(r, s|r_{0}) = \frac{1}{4\pi r_{0}\sqrt{4Ds}} \exp\left\{-|r-r_{0}|\sqrt{\frac{s}{D}}\right\} + \frac{1}{4\pi r_{0}\sqrt{4Ds}} \exp\left\{-(r+r_{0}-2r_{r})\sqrt{\frac{s}{D}}\right\} - \frac{2\left(\frac{1}{r_{r}} + \frac{k_{1}s}{D(s+k_{-1})}\right)}{\left(\frac{1}{r_{r}} + \frac{k_{1}s}{D(s+k_{-1})} + \sqrt{\frac{s}{D}}\right)} \frac{\exp\left\{-(r+r_{0}-2r_{r})\sqrt{\frac{s}{D}}\right\}}{4\pi r_{0}\sqrt{4Ds}} - \frac{Z(s)$$
(51)

Applying the inverse Laplace transform leads to

$$rC(r, s|r_0) = \frac{1}{8\pi r_0 \sqrt{\pi Dt}} \exp\left\{-\frac{(r-r_0)^2}{4Dt}\right\} + \frac{1}{8\pi r_0 \sqrt{\pi Dt}} \exp\left\{-\frac{(r+r_0-2r_r)^2}{4Dt}\right\} - \mathcal{L}^{-1}\left\{Z(s)\right\}.$$
(52)

Due to the complexity of Z(s), we can not derive the closedform expression for its inverse Laplace transform  $f_z(t) = \mathcal{L}^{-1} \{Z(s)\}$ . We employ the Gil-Pelaez theorem [21] for the characteristic function to derive the cumulative distribution function (CDF)  $F_z(t)$  as

$$F_{z}\left(t\right) = \frac{1}{2} - \frac{1}{\pi} \int_{0}^{\infty} \frac{e^{-jwt}\varphi_{Z}^{*}\left(w\right) - e^{jwt}\varphi_{Z}\left(w\right)}{2jw} dw, \quad (53)$$

where  $\varphi_Z(w)$  is given in (9).

Taking the derivative of  $F_{z}(t)$ , we derive the inverse Laplace transform of Z(s) as

$$f_{z}(t) = \frac{1}{2\pi} \int_{0}^{\infty} \left( e^{-jwt} \varphi_{Z}^{*}(w) + e^{jwt} \varphi_{Z}(w) \right) dw.$$
 (54)

Combining (52) and (9), we finally derive the expected timevarying spatial distribution in (8).

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